

Big Data Examination

Roll No. - DS5B-2121

Question 2

Q.II Considering the Item_Outlet_Sales as dependent variable in "Big Mart Sales" dataset, determine the accuracy of the model. Split the dataset according to your last digit of roll no. (Example: if your roll no is ending with 0, the ratio will be 70, 30; if your roll no is ending with 1, the ratio will be 71, 29; if your roll no is ending with 2, the ratio will be 72, 28; if your roll no is ending with 3, the ratio will be 73, 27 etc.).

Importing Pyspark Library

```
In [ ]: pip install pyspark
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting pyspark
  Downloading pyspark-3.2.1.tar.gz (281.4 MB)
    |████████████████████████████████████████| 281.4 MB 32 kB/s
Collecting py4j==0.10.9.3
  Downloading py4j-0.10.9.3-py2.py3-none-any.whl (198 kB)
    |████████████████████████████████████████| 198 kB 69.0 MB/s
Building wheels for collected packages: pyspark
  Building wheel for pyspark (setup.py) ... done
  Created wheel for pyspark: filename=pyspark-3.2.1-py2.py3-none-any.whl size=281853642 sha256=bf2c426d86882346b14a6fc9d4f4fe8f8f6c664e20345f63052db0e57c752d41
  Stored in directory: /root/.cache/pip/wheels/9f/f5/07/7cd8017084dce4e93e84e92efd1e1d5334db05f2e83bcef74f
Successfully built pyspark
Installing collected packages: py4j, pyspark
Successfully installed py4j-0.10.9.3 pyspark-3.2.1
```

Import Library and Creating Session

```
In [ ]: from pyspark.sql import SparkSession
```

```
In [ ]: session = SparkSession.builder.appName("Big_Mart_Sales_Dataset").master("local").getOrCreate()
#we reassign value of __name__ (inbuilt variable) to "__main__" and main is used as entrance
# else the value of name might be different
```

Reading the Data From CSV

```
In [ ]: data = session.read.csv("Big Mart Sale.csv", header = True, inferSchema=True)
```

To print top 10 raw in dataset

```
In [ ]: data.show(10)
```

```
+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+
```

```

+-----+-----+-----+-----+-----+-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
|          FDA15|          9.3|          Low Fat|          0.016047301|          Dairy|249.8092|          OUT049|          1999|          Medium|          Tier 1|Supermarket Type1|          3735.138|
|          DRC01|          5.92|          Regular|          0.019278216|          Soft Drinks| 48.2692|          OUT018|          2009|          Medium|          Tier 3|Supermarket Type2|          443.4228|
|          FDN15|          17.5|          Low Fat|          0.016760075|          Meat| 141.618|          OUT049|          1999|          Medium|          Tier 1|Supermarket Type1|          2097.27|
|          FDX07|          19.2|          Regular|          0.0|Fruits and Vegetables| 182.095|          OUT010|          1998|          null|          Tier 3|Grocery Store|          732.38|
|          NCD19|          8.93|          Low Fat|          0.0|          Household| 53.8614|          OUT013|          1987|          High|          Tier 3|Supermarket Type1|          994.7052|
|          FDP36|          10.395|          Regular|          0.0|          Baking Goods| 51.4008|          OUT018|          2009|          Medium|          Tier 3|Supermarket Type2|          556.6088|
|          FDO10|          13.65|          Regular|          0.012741089|          Snack Foods| 57.6588|          OUT013|          1987|          High|          Tier 3|Supermarket Type1|          343.5528|
|          FDP10|          null|          Low Fat|          0.127469857|          Snack Foods|107.7622|          OUT027|          1985|          Medium|          Tier 3|Supermarket Type3|          4022.7636|
|          FDH17|          16.2|          Regular|          0.016687114|          Frozen Foods| 96.9726|          OUT045|          2002|          null|          Tier 2|Supermarket Type1|          1076.5986|
|          FDU28|          19.2|          Regular|          0.09444959|          Frozen Foods|187.8214|          OUT017|          2007|          null|          Tier 2|Supermarket Type1|          4710.535|
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
only showing top 10 rows

```

Check Null Values in columns

```
In [ ]: from pyspark.sql.functions import isnan, when, count, col
```

```
In [ ]: data.select([count(when(isnan(c) | col(c).isNull(), c)).alias(c) for c in data.columns])
```

```

+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
|          0|          1463|          0|          0|          0|          0|          0|
|          0|          0|          2410|          0|          0|          0|
|          0|
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+

```

Importing sql functions

```
In [ ]: import pyspark.sql.functions as func
```

```
In [ ]: data.agg(func.percentile_approx("Item_Weight", 0.5).alias("mean")).show()
```

```
+-----+
| mean |
+-----+
| 12.6 |
+-----+
```

Fill Null Values

```
In [ ]: # replace 12.6 in place of Null values in Item_weight column because it is mean in this
data = data.na.fill(value=12.6,subset=["Item Weight"])
```

```
In [ ]: # return Medium in place of Null values in Outlet_Size Column Because Medium is the medi
data = data.na.fill(value="Medium", subset=["Outlet Size"])
```

```
In [ ]: data.show()
```

Item and Outlet Information							Sales Performance		
Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_Outlet_Sales	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	Outlet_Location_Type
MRP092	249.8	Low Fat	0.016047301	Dairy	3735.138	OUT049	1999	Medium	Tier 1
Market Type1	3735.138								Supermarket
MRP692	48.2	Regular	0.019278216	Soft Drinks	443.4228	OUT018	2009	Medium	Tier 3
Market Type2	443.4228								Supermarket
MRP618	141.0	Low Fat	0.016760075	Meat	2097.27	OUT049	1999	Medium	Tier 1
Market Type1	2097.27								Supermarket
MRP095	182.0	Regular	0.0	Fruits and Vegetables	732.38	OUT010	1998	Medium	Tier 3
Cery Store	732.38								Grocery
MRP614	53.8	Low Fat	0.0	Household	994.7052	OUT013	1987	High	Tier 3
Market Type1	994.7052								Supermarket
MRP008	51.4	Regular	0.0	Baking Goods	556.6088	OUT018	2009	Medium	Tier 3
Market Type2	556.6088								Supermarket
MRP588	57.6	Regular	0.012741089	Snack Foods	343.5528	OUT013	1987	High	Tier 3
Market Type1	343.5528								Supermarket
MRP622	107.7	Low Fat	0.127469857	Snack Foods	4022.7636	OUT027	1985	Medium	Tier 3
Market Type3	4022.7636								Supermarket
MRP	96.9	Regular	0.016687114	Frozen Foods					

```

726|      OUT045|      1076.5986|      2002|      Medium|      Tier 2|Superma
rket Type1|      1076.5986|      Regular|      0.09444959|      Frozen Foods|187.8
|      FDU28|      19.2|      2007|      Medium|      Tier 2|Superma
214|      OUT017|      4710.535|      Low Fat|      0.0|Fruits and Vegeta...| 45.5
|      FDY07|      11.8|      1999|      Medium|      Tier 1|Superma
402|      OUT049|      1516.0266|      Regular|      0.045463773|      Dairy|144.1
rket Type1|      1516.0266|      1997|      Small|      Tier 1|Superma
|      FDA03|      18.5|      Regular|      0.1000135|Fruits and Vegeta...|145.4
102|      OUT046|      2187.153|      1999|      Medium|      Tier 1|Superma
rket Type1|      2187.153|      Regular|      0.1000135|Fruits and Vegeta...|145.4
|      FDX32|      15.1|      1999|      Medium|      Tier 1|Superma
786|      OUT049|      1589.2646|      Regular|      0.047257328|      Snack Foods|119.6
rket Type1|      1589.2646|      1997|      Small|      Tier 1|Superma
|      FDS46|      17.6|      Regular|      0.0680243|Fruits and Vegeta...|196.4
782|      OUT046|      2145.2076|      Low Fat|      0.0680243|Fruits and Vegeta...|196.4
rket Type1|      2145.2076|      1987|      High|      Tier 3|Superma
|      FDF32|      16.35|      Regular|      0.069088961|      Breakfast| 56.3
426|      OUT013|      1977.426|      1997|      Small|      Tier 1|Superma
rket Type1|      1977.426|      Regular|      0.069088961|      Breakfast| 56.3
|      FDP49|      9.0|      1997|      Small|      Tier 1|Superma
614|      OUT046|      1547.3192|      Low Fat|      0.008596051|      Health and Hygiene|115.3
rket Type1|      1547.3192|      2009|      Medium|      Tier 3|Superma
|      NCB42|      11.8|      Low Fat|      0.008596051|      Health and Hygiene|115.3
492|      OUT018|      1621.8888|      Regular|      0.069196376|      Breakfast| 54.3
rket Type2|      1621.8888|      1999|      Medium|      Tier 1|Superma
|      FDP49|      9.0|      Regular|      0.069196376|      Breakfast| 54.3
614|      OUT049|      718.3982|      Low Fat|      0.034237682|      Hard Drinks|113.2
rket Type1|      718.3982|      1985|      Medium|      Tier 3|Superma
|      DRI11|      12.6|      Low Fat|      0.034237682|      Hard Drinks|113.2
834|      OUT027|      2303.668|      Low Fat|      0.10249212|      Dairy|230.5
rket Type3|      2303.668|      2004|      Small|      Tier 2|Superma
|      FDU02|      13.35|      Low Fat|      0.10249212|      Dairy|230.5
352|      OUT035|      2748.4224|      Low Fat|      0.10249212|      Dairy|230.5
rket Type1|      2748.4224|      Low Fat|      0.10249212|      Dairy|230.5
+-----+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+
only showing top 20 rows

```

Exploration Of Dataset

To print all columns name

```
In [ ]: data.columns
```

```
Out[ ]: ['Item_Identifier',
        'Item_Weight',
        'Item_Fat_Content',
        'Item_Visibility',
        'Item_Type',
        'Item_MRP',
        'Outlet_Identifier',
        'Outlet_Establishment_Year',
        'Outlet_Size',
        'Outlet_Location_Type',
        'Outlet_Type',
        'Item_Outlet_Sales']
```

To count total numbers of rows in dataset

```
In [ ]: data.count()
```

Out []: 8523

```
In [ ]: data.printSchema()
```

```
root
 |-- Item_Identifier: string (nullable = true)
 |-- Item_Weight: double (nullable = false)
 |-- Item_Fat_Content: string (nullable = true)
 |-- Item_Visibility: double (nullable = true)
 |-- Item_Type: string (nullable = true)
 |-- Item_MRP: double (nullable = true)
 |-- Outlet_Identifier: string (nullable = true)
 |-- Outlet_Establishment_Year: integer (nullable = true)
 |-- Outlet_Size: string (nullable = false)
 |-- Outlet_Location_Type: string (nullable = true)
 |-- Outlet_Type: string (nullable = true)
 |-- Item_Outlet_Sales: double (nullable = true)
```

To know data type of each columns

```
In [ ]: data.dtypes
```

```
Out [ ]: [('Item_Identifier', 'string'),
 ('Item_Weight', 'double'),
 ('Item_Fat_Content', 'string'),
 ('Item_Visibility', 'double'),
 ('Item_Type', 'string'),
 ('Item_MRP', 'double'),
 ('Outlet_Identifier', 'string'),
 ('Outlet_Establishment_Year', 'int'),
 ('Outlet_Size', 'string'),
 ('Outlet_Location_Type', 'string'),
 ('Outlet_Type', 'string'),
 ('Item_Outlet_Sales', 'double')]
```

Data Preprocessing

Here we convert the data into machine readable form

```
In [ ]: from pyspark.ml.feature import VectorAssembler, StringIndexer, OneHotEncoder
# It is use for mapping a string column to a index column that will be treated as a cate
# It is feature transformer that combine multiple columns into a single vector column.
# Pyspark ml models takes only one independent variable and one dependent varibale
#but, we have multiple independent variabales, so we use vector assembler to convert the
# of independent variables
```

```
In [ ]: # Doing String Indexing
str_index = StringIndexer(inputCols = ['Item_Identifier','Item_Fat_Content','Item_Type',
```

```
In [ ]: # One Hot Encoding
one_hot = OneHotEncoder(inputCols = ['Item_Identifier1','Item_Fat_Content1','Item_Type1',
```

```
In [ ]: # APply Vector Assembler
vector_ass = VectorAssembler(inputCols = ['Item_Weight','Item_Fat_Content2','Item_Visibi
```

Import Linear Regression and Create Model

```
In [ ]: from pyspark.ml.regression import LinearRegression
```

```
In [ ]: linear = LinearRegression(featuresCol="allfeatures", labelCol="Item_Outlet_Sales")
```

Create Pipeline for ML Model

```
In [ ]: from pyspark.ml import Pipeline
mypipeline = Pipeline(stages = [str_index, one_hot, vector_ass, linear])
```

Making Train Test Split

Splitting the Dataset

As my roll no is DS5B-2121 I will be using split as 0.71 and 0.29

```
In [ ]: training, test = data.randomSplit([0.71, 0.29])
```

Model Training

```
In [ ]: lin_reg_model = mypipeline.fit(training)
```

Test Model

```
In [ ]: result = lin_reg_model.transform(test)
```

```
In [ ]: result.show()
```

+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----														
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-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----														
Item_Identifier Item_Weight Item_Fat_Content Item_Visibility Item_Type Item_MRP Outlet_Identifier Outlet_Establishment_Year Outlet_Size Outlet_Location_Type Outlet_Type Item_Outlet_Sales Item_Identifier1 Item_Fat_Content1 Item_Type1 Outlet_Identifier1 Outlet_Establishment_Year1 Outlet_Size1 Outlet_Location_Type1 Outlet_Type1 Item_Identifier2 Item_Fat_Content2 Item_Type2 Outlet_Identifier2 Outlet_Establishment_Year2 Outlet_Size2 Outlet_Location_Type2 Outlet_Type2 allfeatures prediction														
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-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----														
	DRA12	11.6	Low Fat	0.0	Soft Drinks	141.6154								
	OUT045	2002	Medium		Tier 2	Supermarket Type								
1	3829.0158	521.0	0.0	8.0	1.0									
	2.0	0.0	1.0	0.0	(1543, [521], [1.0])	(4, [0], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2, [0], [1.0])	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 1, 14, 21, 22...	2225.0090020780335
	DRA12	11.6	Low Fat	0.040911824	Soft Drinks	142.3154								

	OUT013	1987	High	Tier 3 Supermarket Type	
1	2552.6772	521.0	0.0	8.0	6.0
	6.0	2.0	0.0	0.0	(1543,[521],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[6],[1.0])	(8,[6],[1.0])	
(2,[],[])	(2,[0],[1.0])	(3,[0],[1.0])	(29,[0,1,5,14,21,...]	2224.685214756014	
	DRA24	19.35	Regular	0.039895009 Soft Drinks	162.4868
	OUT013	1987	High	Tier 3 Supermarket Type	
1	4422.2436	77.0	1.0	8.0	6.0
	6.0	2.0	0.0	0.0	(1543,[77],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[6],[1.0])	(8,[6],[1.0])	
(2,[],[])	(2,[0],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...]	2597.340924578537	
	DRA59	8.27	Regular	0.127927931 Soft Drinks	184.8924
	OUT046	1997	Small	Tier 1 Supermarket Type	
1	4442.2176	78.0	1.0	8.0	2.0
	3.0	1.0	2.0	0.0	(1543,[78],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[2],[1.0])	(8,[3],[1.0])	(2,[1],[1.0])
(2,[],[])	(2,[],[])	(3,[0],[1.0])	(29,[0,2,5,14,21,...]	2982.1821632476695	
	DRA59	12.6	Regular	0.127308434 Soft Drinks	186.6924
	OUT027	1985	Medium	Tier 3 Supermarket Type	
3	7033.5112	78.0	1.0	8.0	4.0
	0.0	0.0	0.0	3.0	(1543,[78],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[4],[1.0])	(8,[0],[1.0])	(2,[0],[1.0])
(2,[],[])	(2,[0],[1.0])	(3,[],[])	(29,[0,2,5,14,21,...]	4331.305955126552	
	DRB01	7.39	Low Fat	0.082170947 Soft Drinks	190.953
	OUT013	1987	High	Tier 3 Supermarket Type	
1	2466.789	1271.0	0.0	8.0	6.0
	6.0	2.0	0.0	0.0	(1543,[1271],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[6],[1.0])	(8,[6],[1.0])	
(2,[],[])	(2,[0],[1.0])	(3,[0],[1.0])	(29,[0,1,5,14,21,...]	2966.2617325300225	
	DRB13	6.115	Regular	0.007055292 Soft Drinks	188.653
	OUT049	1999	Medium	Tier 1 Supermarket Type	
1	3605.307	522.0	1.0	8.0	5.0
	5.0	0.0	2.0	0.0	(1543,[522],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[5],[1.0])	(8,[5],[1.0])	(2,[0],[1.0])
(2,[],[])	(2,[],[])	(3,[0],[1.0])	(29,[0,2,5,14,21,...]	2973.506639206491	
	DRB24	8.785	Low Fat	0.020609218 Soft Drinks	155.1656
	OUT049	1999	Medium	Tier 1 Supermarket Type	
1	4016.1056	906.0	0.0	8.0	5.0
	5.0	0.0	2.0	0.0	(1543,[906],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[5],[1.0])	(8,[5],[1.0])	(2,[0],[1.0])
(2,[],[])	(2,[],[])	(3,[0],[1.0])	(29,[0,1,5,14,21,...]	2420.7896947929494	
	DRB25	12.6	Low Fat	0.06912336 Soft Drinks	106.0938
	OUT027	1985	Medium	Tier 3 Supermarket Type	
3	2787.0388	247.0	0.0	8.0	4.0
	0.0	0.0	0.0	3.0	(1543,[247],[1.0])
0))	(4,[0],[1.0])	(15,[8],[1.0])	(9,[4],[1.0])	(8,[0],[1.0])	(2,[0],[1.0])
(2,[],[])	(2,[0],[1.0])	(3,[],[])	(29,[0,1,5,14,21,...]	3046.1042772127885	
	DRB48	12.6	Regular	0.024733134 Soft Drinks	40.2822
	OUT027	1985	Medium	Tier 3 Supermarket Type	
3	1296.3126	248.0	1.0	8.0	4.0
	0.0	0.0	0.0	3.0	(1543,[248],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[4],[1.0])	(8,[0],[1.0])	(2,[0],[1.0])
(2,[],[])	(2,[0],[1.0])	(3,[],[])	(29,[0,2,5,14,21,...]	2058.027429105631	
	DRB48	16.75	Regular	0.024832806 Soft Drinks	38.7822
	OUT013	1987	High	Tier 3 Supermarket Type	
1	667.7974	248.0	1.0	8.0	6.0
	6.0	2.0	0.0	0.0	(1543,[248],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[6],[1.0])	(8,[6],[1.0])	
(2,[],[])	(2,[0],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...]	656.3620458837856	
	DRC01	5.92	Regular	0.019184026 Soft Drinks	50.3692
	OUT013	1987	High	Tier 3 Supermarket Type	
1	591.2304	907.0	1.0	8.0	6.0
	6.0	2.0	0.0	0.0	(1543,[907],[1.0])
0))	(4,[1],[1.0])	(15,[8],[1.0])	(9,[6],[1.0])	(8,[6],[1.0])	
(2,[],[])	(2,[0],[1.0])	(3,[0],[1.0])	(29,[0,2,5,14,21,...]	806.9476679936189	
	DRC01	5.92	Regular	0.019238942 Soft Drinks	49.8692

```

OUT045|      2002|      Medium|      Tier 2|Supermarket Type
1|      1133.1916|      907.0|      1.0|      8.0|      1.0|
      2.0|      0.0|      1.0|      0.0| (1543,[907],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[1],[1.0])|      (8,[2],[1.0])|(2,
[0],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,2,5,14,21,...| 803.1797886168849|
|      DRC01|      5.92|      Regular|      0.019308607|Soft Drinks| 49.0692|
      OUT017|      2007|      Medium|      Tier 2|Supermarket Type
1|      1478.076|      907.0|      1.0|      8.0|      7.0|
      7.0|      0.0|      1.0|      0.0| (1543,[907],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[7],[1.0])|      (8,[7],[1.0])|(2,
[0],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,2,5,14,21,...| 790.6472831235122|
|      DRC12|      17.85|      Low Fat|      0.037826873|Soft Drinks|189.7188|
      OUT046|      1997|      Small|      Tier 1|Supermarket Type
1|      2285.0256|      908.0|      0.0|      8.0|      2.0|
      3.0|      1.0|      2.0|      0.0| (1543,[908],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[2],[1.0])|      (8,[3],[1.0])|(2,
[1],[1.0])|      (2,[],[])|(3,[0],[1.0])|(29,[0,1,5,14,21,...|3067.6173482995546|
|      DRC25|      5.73|      Low Fat|      0.045334098|Soft Drinks| 87.0882|
      OUT013|      1987|      High|      Tier 3|Supermarket Type
1|      1803.6522|      79.0|      0.0|      8.0|      6.0|
      6.0|      2.0|      0.0|      0.0| (1543,[79],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[6],[1.0])|      (8,[6],[1.0])|
(2,[],[])|      (2,[0],[1.0])|(3,[0],[1.0])|(29,[0,1,5,14,21,...|1342.3755568793158|
|      DRC27|      13.8|      Low Fat|      0.058091482|      Dairy|245.1802|
      OUT035|      2004|      Small|      Tier 2|Supermarket Type
1|      5650.6446|      1272.0|      0.0|      4.0|      3.0|
      4.0|      1.0|      1.0|      0.0| (1543,[1272],[1.
0])|      (4,[0],[1.0])|(15,[4],[1.0])|      (9,[3],[1.0])|      (8,[4],[1.0])|(2,
[1],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,1,5,10,21,...|3999.3506683213764|
|      DRC27|      13.8|      Low Fat|      0.058192802|      Dairy|246.9802|
      OUT049|      1999|      Medium|      Tier 1|Supermarket Type
1|      5896.3248|      1272.0|      0.0|      4.0|      5.0|
      5.0|      0.0|      2.0|      0.0| (1543,[1272],[1.
0])|      (4,[0],[1.0])|(15,[4],[1.0])|      (9,[5],[1.0])|      (8,[5],[1.0])|(2,
[0],[1.0])|      (2,[],[])|(3,[0],[1.0])|(29,[0,1,5,10,21,...|3940.5852931316194|
|      DRC27|      13.8|      Low Fat|      0.058339153|      Dairy|246.2802|
      OUT018|      2009|      Medium|      Tier 3|Supermarket Type
2|      1228.401|      1272.0|      0.0|      4.0|      0.0|
      1.0|      0.0|      0.0|      2.0| (1543,[1272],[1.
0])|      (4,[0],[1.0])|(15,[4],[1.0])|      (9,[0],[1.0])|      (8,[1],[1.0])|(2,
[0],[1.0])|      (2,[0],[1.0])|(3,[2],[1.0])|(29,[0,1,5,10,21,...| 3612.24217303598|
|      DRC27|      13.8|      Low Fat|      0.097251621|      Dairy|245.7802|
      OUT010|      1998|      Medium|      Tier 3|      Grocery Stor
e|      245.6802|      1272.0|      0.0|      4.0|      8.0|
      8.0|      0.0|      0.0|      1.0| (1543,[1272],[1.
0])|      (4,[0],[1.0])|(15,[4],[1.0])|      (9,[8],[1.0])|      (8,[],[])|(2,
[0],[1.0])|      (2,[0],[1.0])|(3,[1],[1.0])|(29,[0,1,5,10,21,...|1972.0395500213303|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

```

Evaluate Model

```
In [ ]: from pyspark.ml.evaluation import RegressionEvaluator
```

```
In [ ]: errors = ["r2", "rmse", "mse", "mae"]
name = ["R-Square or Accuracy", "Root Mean Square Error", "Mean Square Error", "Mean Abs
```



```
for i in range(len(errors)):
    eval = RegressionEvaluator(predictionCol="prediction", labelCol='Item_Outlet_Sales', m
    print("The {} of Model is {}".format(name[i],eval.evaluate(result)))
```

```
The R-Square or Accuracy of Model is 0.5544914690571232
The Root Mean Square Error of Model is 1159.6628884131828
The Mean Square Error of Model is 1344818.0147628062
The Mean Absolute Error of Model is 853.5910257087111
```

In []: